

MYOPIC SEARCH PLANS

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THESIS

MYOPIC SEARCH PLANS

by

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(20. ABSTRACT Continued)

This thesis documents interactive computer programs that are useful for testing search strategies against the myopic strategy, and shows examples where the myopic strategy is not optimal.

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Myopic Search Plans

by

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ABSTRACT

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If the searcher's action at each time unit maximizes his chances of immediate detection, his strategy is said to be myopic. If, however, the searcher seeks to allocate search effort to maximize the probability of detecting the target within a preset amount of time, his strategy is called optimal.

This thesis documents interactive computer programs that are useful for testing search strategies against the myopic strategy, and shows examples where the myopic strategy is not optimal.

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I. INTRODUCTION

A search is an operation with the purpose of finding an object (target) that will in the thesis be assumed to move during the search.

In general, it will be assumed that the target is within an area that can be partitioned into a finite number of cells, moving from one cell to another at fixed time intervals (periods). The actual path that the object follows, whenever it moves, is unknown but the probabilistic distribution of all possible paths is given. It is assumed that the cells which the target will occupy in the future depend only on the cell it is in at the present time but not on how it got there (markovian motion).

If the searcher allocates his resources to maximize the chances of immediate detection, his search plan is said to be myopic. If, however, he seeks to allocate search effort to maximize the probability of detection within a preset amount of time, his plan is called optimal.

In an important class of naval search problems, the target probabilistic distribution over space, i.e., the initial probability of it being in each cell is known. If resources are allocated to some cells, a new distribution over space is to be considered after an unsuccessful search, according to Bayes' theorem.

Formally, it is assumed that the following are given:

- a) A set C of states.
- b) A set A of actions.
- c) An apriori distribution $P_1(c)$ defined on C with $P_1(c)$ being the probability that the target is initially in state c.
- d) A function $Q(a,c)$ being the probability of no detection if action a is taken when target is in state c.
- e) A function $M(c/d)$ being the probability that the state of the target changes from d to c between actions. Since only the current state d is relevant to determine c, the target motion is markovian.

Let a_1, a_2, \dots, a_t be a sequence of actions.

Let $P'_t(c)$ be the probability distribution of the state of the target conditioned on non-detection by a_1, a_2, \dots, a_t and $t-1$ state changes, and $P_{t+1}(c)$ be the probability distribution conditioned on non-detection by a_1, a_2, \dots, a_t and t state changes.

Then, according to Bayes theorem,

$$P'_t(c) = \frac{P_t(c)Q(a_t, c)}{\sum_d P_t(d)Q(a_t, d)} \quad (1)$$

and according to the motion model

$$P_{t+1}(c) = \sum_d P'_t(d)M(c/d) \quad (2)$$

Since $P_1(c)$ is given, alternative applications of (1) and (2) will provide $P_t(c)$ for all t .

The probability of detection during period t , conditioned on earlier failures is

$$P_t = \sum_d P_t(d) [1 - Q(a_t, d)]$$

If a_t is chosen to maximize P_t for $t = 1, 2, \dots$ in succession, then a_t is a myopic plan.

The probability that the target has not been detected until the end of the T^{th} period is

$$\bar{P}(T) = P[\text{target not detected at period 1, and not detected at period 2, and } \dots, \text{ and not detected at period } T]$$

$$\bar{P}(T) = \prod_{t=1}^T (1 - P_t) = \prod_{t=1}^T [\sum_d P_t(d) Q(a_t, d)]$$

Then, the probability of detection after T actions

$$P(T) = 1 - \bar{P}(T) = 1 - \prod_{t=1}^T [\sum_d P_t(d) Q(a_t, d)]$$

a_t is an optimal plan if it maximizes $P(T)$ for a given T .

Optimal plans are highly complex and consequently expensive to find [1]. In contrast, myopic plans are easy to find.

The myopic strategy may be optimal, as in the case of stationary targets, or near optimal, as in most of the examples in references [1] and [2].

There are however, cases for which myopic plans are strongly non-optimal.

Models are to be constructed and used to reach the main goals of this thesis:

- 1) Develop and implement an algorithm to find myopic plans.
- 2) Create interactive search programs, and
- 3) Use them to discover classes of problems for which the myopic strategy is strongly non-optimal.

II. THE MODELS AND ASSUMPTIONS

A. SPACE MODEL

Space is divided into $m \times n$ square cells, each one identified by a 2-tuple. The upper left cell is cell (1,1). Cell (i,j) is the i^{th} cell to the east, j^{th} cell to the south.

If it happens that a target cannot move in some directions, boundaries can be introduced, either reflecting or absorbing. A target cannot cross a reflecting boundary. If one of its paths leads to the outside, this path is reflected, i.e., the target moves in the opposite direction. Reflecting boundaries model, for example, the borders of a channel. An absorbing boundary can be crossed from the interior but not from outside. It models the case in which the target has some information about the search area and tries to evade. Once it is out, it will never move back into that area.

By search area it is meant the subset of cells to which the searcher is able, allowed or willing to allocate effort and that is not necessarily the same subset that the target can move across. The latter will be referred to as the target area.

B. DETECTION MODEL

Sensors are assumed to have an exponential detection function, that is, the conditional probability of detection has the form:

$$1 - \exp[-a(g,t)x(g,t)],$$

where $x(g,t)$ is the amount of search effort allocated to cell g at period t and $a(g,t)$ is a non-negative constant which may depend on the cell, may change with time, and that will be referred to as detection rate.

C. THE MOTION MODELS

Two models are used: the random walk in space and the random walk in speed.

In both models, speed is expressed in terms of cells per period.

If a target occupies cell (i,j) and, after a change of state it is in cell $(i+k,j+l)$, its speed in the west-east direction is $V_x = k$ cells/period, and its speed in the north-south direction is $V_y = l$ cells/period

1. Random Walk in Space

In this model, $P_t(c) = S_t(i,j)$, i.e., $P_t(c)$ is the probability of the target being in cell (i,j) .

Given the joint distribution of V_x and V_y , $t_{V_x, V_y}(v_x, v_y)$, invariant with time and space,

$$S_{t+1}(i,j) = \sum_k \sum_l S_t(k,l) t_{V_x, V_y}(i-k, j-l)$$

The, for this model,

$$M(c/d) = t_{V_x, V_y}(i-k, j-l)$$

where i and j define state c , and the state d is defined by k and ℓ .

2. Random Walk in Speed

Sets of possible values for V_x and V_y are fixed.

Let $V^x = \{v_1^x, v_2^x, \dots, v_n^x\}$ and $V^y = \{v_1^y, v_2^y, \dots, v_m^y\}$

be the sets of possible values that V_x and V_y can assume.

$P_t(c)$ is equal to $S_t(i, j, k, \ell)$, i.e., $P_t(c)$ is the probability of the target being in cell (i, j) , its speed having components, v_k^x, v_ℓ^y .

Let $P_{\Delta V_x}(\delta V_x)$ and $P_{\Delta V_y}(\delta V_y)$ be the known discrete distributions of ΔV_x and ΔV_y , the changes in V_x and V_y per period, respectively.

Given $P_{\Delta V_x}(\delta V_x)$ a matrix P^x can be constructed, which entries $p_{i,j}^x$ are the probabilities of V_x changing from v_i^x to v_j^x in one period. A similar matrix P^y can be constructed which entries are the probabilities of the changes in V_y . Then

$$S_{t+1}(i, j, k, \ell) = \sum_r \sum_s S_t(i-r, j-s, r, s) p_{r,k}^x p_{s,\ell}^y.$$

III. MYOPIC SEARCH PLANS

Let p_i , $i = 1, 2, \dots, n$ be the probability of the target being in the i^{th} of the n cells among which the search effort is to be myopically distributed at period t .

Let x_i , $i = 1, 2, \dots, n$ be the fraction of effort allocated to each of the cells that are assumed to have a common detection rate a which may change with periods.

Given the detection model, a myopic plan maximizes

$$\sum_i p_i (1 - e^{-ax_i}) , \quad i = 1, 2, \dots, n$$

for each $t = 1, 2, \dots$ in succession.

The sum of all x_i must not exceed X , the total amount of effort available in the period and no x_i can be less than 0.

Thus, a myopic plan is the solutions of a sequence of non-linear programs with the form

$$\text{Min } \sum_i p_i e^{-ax_i} \quad (1)$$

$$\text{S/T } \sum_i x_i \leq X \quad (2)$$

$$x_i \geq 0 , \quad i = 1, 2, \dots, n \quad (3)$$

It can be easily proved that equality holds for (2) at optimality.

If the constraints (3) are relaxed and Lagrange method is used:

$$L(x, \lambda) = \sum_i p_i e^{-ax_i} + \lambda (\sum_i x_i - X)$$

$$\frac{\partial L}{\partial x_i} = -ap_i e^{-ax_i} = 0 \quad (4)$$

$$\frac{\partial L}{\partial \lambda} = \sum_i x_i - X = 0 \quad (5)$$

From (4)

$$\lambda = ap_i e^{-ax_i}$$

$$\ln \lambda = \ln(ap_i) - ax_i \quad (6)$$

$$x_i = \frac{\ln a}{a} + \frac{\ln p_i}{a} - \frac{\ln \lambda}{a} \quad (7)$$

Sum (6), side by side, over all i :

$$n \ln \lambda = \sum_i \ln(ap_i) - a \sum_i x_i \quad (8)$$

Substitute X for $\sum_i x_i$ in (8) and rearrange:

$$\ln \lambda = \ln a + \frac{1}{n} \ln[\prod_i p_i] - \frac{aX}{n} \quad (9)$$

Substitute (9) for $\ln \lambda$ in (7):

$$\begin{aligned} x_i^* &= \frac{\ln a}{a} + \frac{\ln p_i}{a} - \frac{1}{a} [\ln a + \frac{1}{n} \ln(\prod_i p_i) - \frac{aX}{n}] \\ x_i^* &= \frac{\ln p_i}{a} - \frac{1}{an} \ln[\prod_i p_i] + \frac{X}{n}, \quad i = 1, 2, \dots, n \end{aligned} \quad (10)$$

Brown [2] proved that the objective function is convex. Then (10) is the optimal solution of the N.L.P. (1), (2). At optimality,

$$p_i e^{-ax_i} = \frac{\lambda}{a} = \text{constant for } i = 1, 2, \dots, n.$$

A. ALGORITHM FOR FINDING MYOPIC PLANS

Provided the detection rate is invariant with cells:

Step 1) Let $I = \{i: i = 1, 2, \dots, n\}$ be the set of indexes of all cells among which the search effort is to be myopically distributed.

Step 2) Solve the N.L.P. (1), (2):

$$x_i = \frac{\ln p_i}{a} + \frac{X}{n} - \frac{\ln P}{an}, \quad i \in I$$

where $P = \prod_{i \in I} p_i$ and m is the number of elements in I .

If $x_i \geq$ for all $i \in I$, stop.

Step 3) Select the cell with smallest p_i , $i \in I$. Remove its index from I and make $x_i = 0$
Go to step 2.

At optimality, $p_j \leq p_i e^{-ax_i^*} = \text{constant}$, for all $j \notin I$,
all $i \in I$.

The algorithm has at most n interactions. Its solution is feasible for the N.L.P. (1), (2), (3) since

$$x_i^* = 0 \quad i \notin I$$

$$x_i^* \geq 0 \quad i \in I$$

$$\sum_i x_i = X$$

and is also optimal.

$$\text{Let } Z^* = p_1 + p_2 + \dots + p_j + p_{j+1} e^{-ax_{j+1}} + \dots + p_{j+m} e^{-ax_{j+m}}$$

where $j+m = n$, be the optimal solution produced by the algorithm after reindexing the cells such that $p_i < p_{i+1}$ for all i .

Then, the algorithm found

$$x_j = \frac{\ln p_j}{a} + \frac{X}{m+1} - \frac{\ln P'}{a(m+1)} \leq 0 \quad (11)$$

where

$$P' = \prod_{i=j}^{j+m} p_i .$$

Suppose the optimal solution was

$$\begin{aligned} \hat{Z}^* = & p_1 + p_2 + \dots + p_{j-1} + p_j e^{-ax'_j} + p_{j+1} e^{-ax'_{j+1}} \\ & + \dots + p_{j+m} e^{-ax'_{j+m}} . \end{aligned}$$

In this case,

$$p_j e^{-ax'_j} = p_{j+1} e^{-ax'_{j+1}} = \dots = p_{j+m} e^{-ax'_{j+m}}$$

and

$$\ln p_j - ax'_j = \ln p_{j+1} - ax'_{j+1}$$

$$\ln p_j - ax'_j = \ln p_{j+2} - ax'_{j+2}$$

$$\vdots$$

$$\ln p_j - ax'_j = \ln p_{j+m} - ax'_{j+m}$$

Sum side by side:

$$m \ln p_j - amx'_j = \sum_{i=j+1}^{j+m} \ln p_i - a \sum_{i=j+1}^{j+m} x'_i$$

$$m \ln p_j - a(m+1)x'_j = \ln \left(\prod_{i=j+1}^{j+m} p_i \right) - aX$$

$$x'_j = \frac{m \ln p_j}{a(m+1)} + \frac{X}{(m+1)} - \frac{\ln P}{a(m+1)} > 0 \quad (12)$$

where

$$P = \prod_{i=j+1}^{j+m} p_i$$

Compare (11) and (12):

$$\frac{m \ln p_j}{a(m+1)} + \frac{X}{(m+1)} - \frac{\ln P}{a(m+1)} > \frac{\ln p_j}{a} + \frac{X}{(m+1)} - \frac{\ln P'}{a(m+1)}$$

Since $a(m+1) > 0$

$$m \ln p_j - \ln P > (m+1) \ln p_j - \ln P'$$

$$\ln P' - \ln P > \ln p_j$$

$$\ln \frac{P'}{P} > \ln p_j$$

$$\ln p_j > \ln p_j \quad (13)$$

Since no x_i can be made 0 before x_{i-1} , (13) proves by contradiction that the solution is optimal.

IV. THE COMPUTER PROGRAMS

Two FORTRAN programs were written.

Program SRCH1 refers to the random walk in space model and allows the space to be divided into a maximum of 100×100 cells. Any subset of cells can be used as target and/or search areas.

In this program, a two dimensional array is defined by the FORTRAN name CELL, which entries are the probabilities of the target being in each of the cells identified by the array indexes.

Program SRCH2 refers to the random walk in speed model. A $25 \times 25 \times 4 \times 4$ array is defined by the FORTRAN name CELL. The entries of this array are the probabilities of the target being in the cells identified by the first two indexes, its speed having the components defined by the last two indexes. Thus, this program allows the space to be divided into a maximum of 25×25 cells and allows the sets of possible values of V_x and V_y to have at most 4 elements each. Any subset of cells can be used as target and/or search areas.

SRCH2 maps the four dimensional array into a two dimensional array called TCELL, which entries have the same meaning as the entries of the array CELL in program SRCH1.

These two dimensional arrays are printed at the beginning of each period, after their entries are coded as follows:

The highest probability is mapped onto 100; the smallest non-zero probability mapped onto 0.

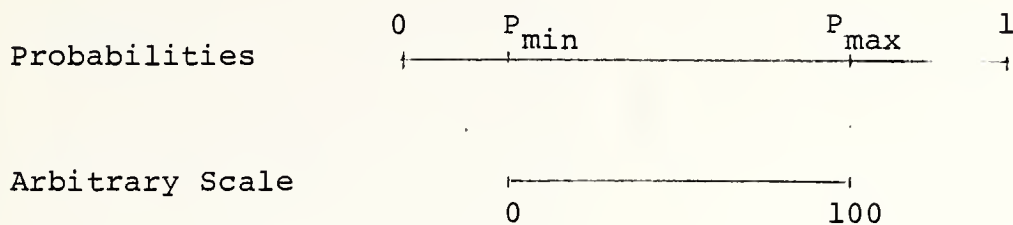


Fig. IV-1

The interval $[0,100]$ is then divided and coded.

Subinterval	Code
100	*
$[99,100)$	9
$[96,99)$	8
$[91,96)$	7
$[84,91)$	6
$[75,84)$	5
$[64,75)$	4
$[51,64)$	3
$[36,51)$	2
$[19,36)$	1
$[0,19)$	0

Probabilities equal 0 are coded as a dot. Fig. IV-2 is an example of the coded distribution of the target.

RANGE OF PROBABILITIES 0.29119E-07 0.17087E-01

30

40

50

0123456789012345678901234

```

27/ 00000000000000000000000000000000
28/ 00000000000000000000000000000000
29/ 00000000111111111100000000000000
30/ 00000011222333322211000000000000
31/ 00000112233333333221100000000000
32/ 00000122344444443221000000000000
33/ 00000123456777654321000000000000
34/ 000000124578*875421000000000000
35/ 000000012345554321000000000000
36/ .000000011222221100000000.
37/ ..000000000000000000000000..
38/ ...0000000000000000000000...
39/ ....00000000000000000000....
40/ .....0000000000000000.....

```

Fig. IV-2

Both programs have a subroutine that distributes myopically the search effort, provided the detection rate is invariant with space within the search area.

Instructions on use of the programs constitute Appendices A and B.

V. EXAMPLES AND CONCLUSIONS

A. EXAMPLES

Different situations were analysed and some of them are now presented.

1. First Example

The a priori distribution of a target was uniform over a 3×3 cell area as follows:

	30	31	32
30	1/9	1/9	1/9
31	1/9	1/9	1/9
32	1/9	1/9	1/9

The search areas was coincident with the above area and the target area was the entire space.

The target moved as follows:

V_x	V_y	$P(V_x=v_x, V_y=v_y)$
1	0	0.25
1	1	0.50
0	1	0.25

Five units of search effort were available and the detection rate was one for the cells within the search area, both values invariant with time.

A myopic and an alternative strategy were used which distributed effort as follows:

PERIOD 1

CELL	STRATEGY	
	MYOPIC	ALTERNATIVE
(30,30)	0.5556	-
(31,30)	0.5556	-
(32,30)	0.5556	1.0
(30,31)	0.5556	-
(31,31)	0.5556	-
(32,31)	0.5556	1.0
(30,32)	0.5556	1.0
(31,32)	0.5556	1.0
(32,32)	0.5556	1.0

PERIOD 2

CELL	STRATEGY	
	MYOPIC	ALTERNATIVE
(32,30)	-	0.313
(31,31)	1.25	-
(32,31)	1.25	1.527
(30,32)	-	0.313
(31,32)	1.25	1.527
(32,32)	1.25	1.32

PERIOD 3

STRATEGY

CELL	MYOPIC	ALTERNATIVE
(31,31)	0.6781	-
(32,31)	1.4071	1.567
(31,32)	1.4071	1.526
(32,32)	1.5076	1.907

PERIOD 4

STRATEGY

CELL	MYOPIC	ALTERNATIVE
(32,31)	1.4331	1.4241
(31,32)	1.4331	1.3205
(32,32)	2.1337	2.2554

PERIOD 5

STRATEGY

CELL	MYOPIC	ALTERNATIVE
(32,32)	5.0	5.0

The probabilities of detection were, after each period were:

STRATEGY		
PERIOD	MYOPIC	ALTERNATIVE
1	0.42624	0.35118
2	0.60819	0.56827
3	0.65128	0.69634
4	0.65790	0.71282
5	0.65814	0.71338

After 5 periods the alternative plan yielded a probability of detection 8.39% higher. This alternative strategy consisted in distributing myopically the effort among the cells adjacent to the boundaries of the search area in the direction of the movement of the target. It made a barrier that the target had to cross to leave the search area.

This example suggested that myopic strategy was strongly non-optimal for cases in which the target could move to a safe region, and the search lasted long enough for it to reach this region.

This example led to the hypothesis that myopic plans were also strongly non-optimal in situations where the target moved to areas where the conditional probability of detection was a strong function of space, the aforementioned example being the extreme case of conditional probability of detection equal zero. However, for other analysed problems, in which the target could also evade to a safe area, no strategies could be found that led to a probability of detection

as much as 4% higher than the probability produced by myopic strategies.

2. Second Example

The a priori distribution of a target was uniform over a 5 x 4 cell area as follows:

	40	41	42	43	44
22	0.05	0.05	0.05	0.05	0.05
23	0.05	0.05	0.05	0.05	0.05
24	0.05	0.05	0.05	0.05	0.05
25	0.05	0.05	0.05	0.05	0.05

The target area was the subset of cells (i,j) such that $38 \leq i \leq 40$, and the search area was such that $j \leq 28$.

The target moved as follows:

v_x	v_y	$P(V_x=v_x, V_y=v_y)$
-1	1	0.3
0	1	0.4
1	1	0.3

Five units of search effort were available and the detection rate was one for the cells within the search area, both values invariant with time.

The interesting point about this case is that four different strategies turned out to be better than the myopic for a seven peeriod search.

The strategies distributed effort as follows:

PERIOD 1					
		STRATEGY			
CELL	MYOPIC	1	2	3	4
(40,22)	0.2500	-	-	-	0.2500
(41,22)	0.2500	-	-	-	0.2500
(42,22)	0.2500	-	-	-	0.2500
(43,22)	0.2500	-	-	-	0.2500
(44,22)	0.2500	-	-	-	0.2500
(40,23)	0.2500	-	-	-	0.2500
(41,23)	0.2500	-	-	-	0.2500
(42,23)	0.2500	-	-	-	0.2500
(43,23)	0.2500	-	-	-	0.2500
(44,23)	0.2500	-	-	-	0.2500
(40,24)	0.2500	-	-	-	0.2500
(41,24)	0.2500	-	-	-	0.2500
(42,24)	0.2500	-	-	-	0.2500
(43,24)	0.2500	-	-	-	0.2500
(44,24)	0.2500	-	-	-	0.2500
(40,25)	0.2500	1.0000	1.0000	1.0000	0.2500
(41,25)	0.2500	1.0000	1.0000	1.0000	0.2500
(42,25)	0.2500	1.0000	1.0000	1.0000	0.2500
(43,25)	0.2500	1.0000	1.0000	1.0000	0.2500
(44,25)	0.2500	1.0000	1.0000	1.0000	0.2500

PERIOD 2

CELL	MYOPIC	STRATEGY			
		1	2	3	4
(40,23)	0.0360	-	-	-	0.0360
(41,23)	0.3927	-	-	-	0.3927
(42,23)	0.3927	-	-	-	0.3927
(43,23)	0.3927	-	-	-	0.3927
(44,23)	0.0360	-	-	-	0.0360
(40,24)	0.0360	-	-	-	0.0360
(41,24)	0.3927	-	-	-	0.3927
(42,24)	0.3927	-	-	-	0.3927
(43,24)	0.3927	-	-	-	0.3927
(44,24)	0.0360	-	-	-	0.0360
(40,25)	0.0360	0.7325	0.7325	0.7325	0.0360
(41,25)	0.3927	1.0892	1.0892	1.0892	0.3927
(42,25)	0.3927	1.0892	1.0892	1.0892	0.3927
(43,25)	0.3927	1.0892	1.0892	1.0892	0.3927
(44,25)	0.0360	0.7325	0.7325	0.7325	0.0360
(40,26)	0.0360				0.0360
(41,26)	0.3927	0.0892	0.0892	0.0892	0.3927
(42,26)	0.3927	0.0892	0.0892	0.0892	0.3927
(43,26)	0.3927	0.0892	0.0892	0.0892	0.3927
(44,26)	0.0360				0.0360

PERIOD 3

STRATEGY

CELL	MYOPIC	1	2	3	4
(40,24)	0.1406				0.1406
(41,24)	0.3229				0.3229
(42,24)	0.3229				0.3229
(43,24)	0.3229				0.3229
(44,24)	0.1406				0.1406
(39,25)		0.0304	0.0304	0.0304	
(40,25)	0.1406	0.7386	0.7386	0.7386	0.1406
(41,25)	0.3229	1.0448	1.0448	1.0448	0.3229
(42,25)	0.3229	1.1391	1.1391	1.1391	0.3229
(43,25)	0.3229	1.0448	1.0448	1.0448	0.3229
(44,25)	0.1406	0.7386	0.7386	0.7386	0.1406
(45,25)		0.0304	0.0304	0.0304	
(40,26)	0.1406	0.0168	0.0168	0.0168	0.1406
(41,26)	0.3229	0.0499	0.0499	0.0499	0.3229
(42,26)	0.3229	0.0499	0.0499	0.0499	0.3229
(43,26)	0.3229	0.0499	0.0499	0.0499	0.3229
(44,26)	0.1406	0.0168	0.0168	0.0168	0.1406
(40,27)	0.1406				0.1406
(41,27)	0.3229				0.3229
(42,27)	0.3229	0.0499	0.0499	0.0499	0.3229
(43,27)	0.3229				0.3229
(44,27)	0.1406				0.1406

PERIOD 4

CELL	MYOPIC	STRATEGY			
		1	2	3	4
(39,25)		0.2006	-	-	-
(40,25)	0.1854	0.7037	-	-	-
(41,25)	0.2930	1.0049	-	-	-
(42,25)	0.2930	1.0944	-	-	-
(43,25)	0.2930	1.0049	-	-	-
(44,25)	0.1854	0.7037	-	-	-
(45,25)	-	0.2006	-	-	-
(39,26)	-	-	0.2259	-	-
(40,26)	0.1854	0.0109	0.3666	-	-
(41,26)	0.2930	0.0109	0.3666	-	-
(42,26)	0.2930	0.0109	0.3666	-	-
(43,26)	0.2930	0.0109	0.3666	-	-
(44,26)	0.1854	0.0109	0.3666	-	-
(45,26)	-	-	0.2259	-	-
(39,27)	-	-	0.0720	0.2624	-
(40,27)	0.1854	-	0.2691	0.4594	-
(41,27)	0.2930	0.0109	0.3666	0.5570	-
(42,27)	0.2930	0.0109	0.3666	0.5570	-
(43,27)	0.2930	0.0109	0.3666	0.5570	-
(44,27)	0.1854	-	0.2691	0.4595	-
(45,27)	-	-	0.0720	0.2624	-
(39,28)	-	-	-	-	0.4636
(40,28)	0.1854	-	0.0144	0.2048	0.7500
(41,28)	0.2930	-	0.2756	0.4661	0.8576
(42,28)	0.2930	-	0.3528	0.5432	0.8576
(43,28)	0.2930	-	0.2756	0.4661	0.8576
(44,28)	0.1854	-	0.0144	0.2048	0.7500
(45,28)	-	-	-	-	0.4636

PERIOD 5

STRATEGY

CELL	MYOPIC	1	2	3	4
(39,26)	0.1551	0.2487	0.3106	-	-
(40,26)	0.2537	0.2939	0.7002	-	-
(41,26)	0.2829	0.2939	0.9634	-	-
(42,26)	0.2829	0.2939	1.0515	-	-
(43,26)	0.2829	0.2939	0.9634	-	-
(44,26)	0.2537	0.2939	0.7002	-	-
(45,26)	0.1551	0.2487	0.3106	-	-
(39,27)	0.1551	0.1902	-	0.5414	-
(40,27)	0.2537	0.2567	-	0.6122	-
(41,27)	0.2829	0.2939	-	0.6524	-
(42,27)	0.2829	0.2939	-	0.6524	-
(43,27)	0.2829	0.2939	-	0.6524	-
(44,27)	0.2537	0.2567	-	0.6122	-
(45,27)	0.1551	0.1902	-	0.5414	-
(39,28)	0.1551	0.0326	-	0.1294	0.5131
(40,28)	0.2537	0.1804	-	0.0953	0.7177
(41,28)	0.2829	0.2687	-	0.0953	0.8357
(42,28)	0.2829	0.2939	-	0.0953	0.8668
(43,28)	0.2829	0.2687	-	0.0953	0.8357
(44,28)	0.2537	0.1804	-	0.0953	0.7177
(45,28)	0.1551	0.0326	-	0.1294	0.5131

PERIOD 6

CELL	MYOPIC	STRATEGY			
		1	2	3	4
(39,27)	0.3795	0.4157	0.5429	0.3880	-
(40,27)	0.3482	0.3398	0.5177	0.6928	-
(41,27)	0.3482	0.3398	0.5177	0.9198	-
(42,27)	0.3482	0.3398	0.5177	0.9987	-
(43,27)	0.3482	0.3398	0.5177	0.9198	-
(44,27)	0.3482	0.3398	0.5177	0.6928	-
(45,27)	0.3795	0.4157	0.5429	0.3880	-
(39,28)	0.3795	0.3851	0.2164	-	0.5516
(40,28)	0.3482	0.3398	0.1887	-	0.7049
(41,28)	0.3482	0.3398	0.1717	-	0.8167
(42,28)	0.3482	0.3398	0.1717	-	0.8535
(43,28)	0.3482	0.3398	0.1717	-	0.8167
(44,28)	0.3482	0.3398	0.1887	-	0.7049
(45,28)	0.3795	0.3851	0.2164	-	0.5516

PERIOD 7

CELL	MYOPIC	STRATEGY			
		1	2	3	4
(38,28)	0.1225	0.1376	0.1727	0.0748	
(39,28)	0.7748	0.7845	0.8073	0.7448	0.5815
(40,28)	0.6411	0.6312	0.6080	0.6721	0.7013
(41,28)	0.6411	0.6312	0.6080	0.6721	0.7996
(42,28)	0.6411	0.6312	0.6080	0.6721	0.8352
(43,28)	0.6411	0.6312	0.6080	0.6721	0.7996
(44,28)	0.6411	0.6312	0.6080	0.6721	0.7013
(45,28)	0.7748	0.7845	0.8073	0.7448	0.5815
(46,28)	0.1225	0.1376	0.1727	0.0728	

The probabilities of detection after each period were:

PERIOD	STRATEGY				
	MYOPIC	1	2	3	4
1	0.22119	0.15803	0.15803	0.15803	0.22119
2	0.38064	0.29861	0.29861	0.29861	0.38064
3	0.49071	0.43139	0.43139	0.43139	0.49071
4	0.57188	0.55845	0.49688	0.48977	0.55116
5	0.63270	0.62597	0.62012	0.54531	0.61063
6	0.67812	0.67652	0.67891	0.66582	0.66948
7	0.70635	0.70815	0.71475	0.72319	0.72790

Strategy 4 was 3.05% better than the myopic, for a seven period search. This strategy searched myopically for the target until it could possibly be in cells adjacent to the boundaries of search area, and then, as in the first example, concentrated the effort in those cells.

B. CONCLUSIONS

Although myopic strategy is strongly non-optimal for some specific cases, as the first example in this thesis and as the problem which Brown [2] called the Island Passage Problem, no classes of problems could be characterized for which a strategy could be found that was much better than the myopic.

Many researched problems, the presented examples inclusive, show that a strategy which may not be optimal but that concentrates the effort near the boundaries of the search area when

the target reaches these boundaries, produces better results than a myopic strategy.

It should be noted that no cases were considered in which the detection rate changed with cells within the search area and none of the alternative strategies used can be guaranteed to be optimal.

C. EXTENSIONS

Extensions can be brought into this thesis that may possibly lead to the characterization of classes of problems for which the myopic strategy is strongly non-optimal.

First, an algorithm to find optimal plans may be implemented and added to the computer programs.

Second, the restriction on the change of detection rate with cells within the search area may be removed from the myopic plan.

Further, other motion model as the fleeing datum [1] and the geometric memory motion [2] can be used.

Last, a detection model in which the conditional probability of detection is a function of the speed of the target can be constructed and used together with the random walk in speed model.

APPENDIX A

INSTRUCTIONS ON USE OF PROGRAM SRCH1

The first data the program requires are the limits of the target area and types of boundaries. They must have the form

aadwwwbbbxxxcccyyydddzzz.

QUANTITY	MEANING	REQUIREMENTS
<u>aaa</u>	West limit	001< <u>aaa</u> <100
<u>bbb</u>	East limit	001< <u>bbb</u> <100; <u>bbb</u> > <u>aaa</u>
<u>ccc</u>	North limit	001< <u>ccc</u> <100
<u>ddd</u>	South limit	001< <u>ddd</u> <100; <u>ddd</u> > <u>ccc</u>
<u>ww</u>	Type of the	
<u>xxx</u>	boundary which	REF for reflecting
<u>yyy</u>	preceeds each of	ABS for absorbing
<u>zzz</u>	these quantities	

Next, the program asks for the limits of the search area.

The entry must have the form

aaabbbcccddd

all quantities with the same meanings and fulfilling the same requirements as above.

Then, the transition matrix is to be introduced.

If the probability of the target moving from cell (m,n) to cell (m+k,n+l) is p_i , enter

aaabbb p

where

aaa = k

bbb = l

p = p_i

enter one line for each i, after which, enter \emptyset .

$\sum_i p_i$ must be equal to 1.

p was the format F10.8

The last entry before calculations begin is the a priori distribution of the target, which must have the form,

eeeffff p

where eee and fff identify the cell where the target is with probability p $\neq \emptyset$.

Enter one line for each p, after which enter \emptyset . p must fulfill the requirements previously stated. From this point on, all actions are repeated at each period.

The program prints a table similar to that in Fig. IV-2 which shows the coded distribution of the target at the beginning of the period, and then, asks for the total amount of effort available in the period and the detection rate. Use format 2F5.3 to introduce these values.

In case the user does not want a myopic distribution, the effort is to be distributed as follows:

eeefff t

where eee and fff identify the cell where the amount t is placed.

Enter one line for each t $\neq 0$.

Summation of all t must add up to the total amount of effort available in the period.

t has the format F10.8.

Enter \emptyset after distributing the search effort.

Questions must be answered Yes (Y) or No (N) as in the following example:

```
$ srchl
EXECUTION BEGINS...
ENTER LIMITS OF TARGET AREA AND TYPE OF BOUNDARIES
>025ref030ref020abs025ref
ENTER LIMITS OF SEARCH AREA
>025030001100
ENTER TRANSITION MATRIX
>000-010.25
>0000010.25
>0010000.25
>-010000.25
>0
ENTER A PRIORI DISTRIBUTION OF TARGET
>0260220.3
>0260230.4
>0270230.3
>0
RANGE OF PROBABILITIES  0.30000    0.40000
    26
    67
    --
22/ 0.
23/ *0
ENTER TOTAL EFFORT AND DETECTION RATE
>3.0001.000
MYOPIC PLAN?
>Y
```


WANT TO KNOW DISTRIBUTION OF EFFORT?

>Y

CELL	EFFORT
26 22	0.9041
26 23	1.1918
27 23	0.9041

AFTER 1 PERIODS, PROB DET IS 0.63559

WANT TO CONTINUE?

>Y

RANGE OF PROBABILITIES 0.833333\$-01 0.16667

25

5678

21/ .0..

22/ 00*.

23/ 0*00

24/ .00.

ENTER TOTAL EFFORT AND DETECTION RATE

>3.0001.000

MYOPIC PLAN?

>n

ENTER DISTRIBUTION OF SEARCH EFFORT

>0270221.5

>0260231.5

>0

AFTER 2 PERIODS, PROB DET IS 0.72995

WANT TO CONTINUE?

>n

WANT TO PLAY AGAIN?

>n

R; T=4.73/9.44 13.30.46

>

APPENDIX B

INSTRUCTIONS ON USE OF PROGRAM SRCH2

The first data the program requires are the limits of the target area and types of boundaries. They must have the form

aawwwbbxxxccyyyddzzz

QUANTITY	MEANING	REQUIREMENTS
<u>aa</u>	West limit	$01 < \underline{aa} < 25$
<u>bb</u>	East limit	$01 < \underline{bb} < 25; \underline{bb} > \underline{aa}$
<u>cc</u>	North limit	$01 < \underline{cc} < 25$
<u>dd</u>	South limit	$01 < \underline{dd} < 25; \underline{dd} > \underline{cc}$
<u>www</u>	Type of the	
<u>xxx</u>	boundary which	REF for reflecting
<u>yyy</u>	preceeds each of	ABS for absorbing
<u>zzz</u>	these quantities	

Next, the program asks for the limits of the search area.

The entry must have the form

aabbccdd

All quantities with the same meanings and fulfilling the same requirements as above.

Then, the program asks how many different values V_x can assume. Any positive integer less than 5 can be introduced, according to Format I2.

Use the same format to introduce the values of V_x , one per line.

The same question is made with respect to V_y , and the same instructions used for V_x apply.

Next, the transition matrix of V_x is to be introduced.

Type

ee p

where ee is the value ΔV_x can assume with probability p.

Enter one line for each p, after which, enter \emptyset .

Summation of all p must be 1.

p has the format F5.3.

The transition matrix for V_y is introduced in accordance with the same rules stated for V_x .

The last data before calculations begin is the a priori distribution of target, which must have the form

ffgg p q₁ q₂ ... r₁ r₂ ...

where ff and gg identify the cell where the target is with probability p.

q₁ q₂ ... are the conditional distribution of V_x , i.e., the probabilities of V_x being v_1^x , v_2^x , ... given the target is in cell (ff,gg). The number of quantities q must be equal to the number of possible values that V_x can assume.

The summation of all q must be 1. The quantities r are the conditional distribution of V_y .

The format for quantities p, q and r is F5.3.

Enter one line for each p $\neq 0$, after which enter \emptyset .

Summation of all p must be one, too.

From this point on, all actions are repeated at each period.

The program prints a table similar to that in Fig. IV-2 which shows the coded distribution of the target at the beginning of the period, and then, asks for the total amount of effort available in the period and the detection rate.

Use format 2F5.3 to introduce these values.

In case the user does not want a myopic distribution, the effort is to be distributed as follows:

ffgg t

where ff and gg identify the cell where the amount t is placed.

Enter one line for each $t \neq 0$.

Summation of all t must add up to the total amount of effort available in the period.

t has the format F10.8

Enter \emptyset after distributing the effort

Questions must be answered Yes (Y) or No (N) as in the following example:

```
>$ srch2
EXECUTION BEGINS...
  ENTER LIMITS OF TARGET AREA AND TYPES OF BOUNDARIES
>l0refl5ref0lref25ref
  ENTER LIMITS OF SEARCH AREA
>01250125
  HOW MANY VALUES CAN V(X) ASSUME?
>03
  ENTER THESE VALUES, ONE PER LINE
>-1
>00
>01
  HOW MANY VALUES CAN V(Y) ASSUME?
>04
  ENTER THESE VALUES, ONE PER LINE
>-1
>00
>01
>02
```



```

ENTER TRANSITION OF V(X)
>-10.3
>000.4
>010.3
>0
ENTER TRANSITION OF V(Y)
>010.8
>020.2
>0
ENTER A PRIORI DISTRIBUTIONS
>12120.3000.4000.3000.3000.1000.2000.3000.400
>12130.4000.5000.2000.3000.6000.2000.1000.100
>13140.3000.4000.4000.2000.5000.3000.1000.100
>0
RANGE OF PROBABILITIES  0.30000      0.4000
    12
    23
    --
12/ 0.
13/ *.
14/ .0
ENTER TOTAL EFFORT AND DETECTION RATE
>1.0001.000
MYOPIC PLAN?
>Y
WANT TO KNOW DISTRIBUTION OF EFFORT?
>Y
    CELL      EFFORT
    12  12    0.2374
    12  13    0.5251
    13  14    0.2374
AFTER 1 PERIODS, PROB DET IS 0.2902
WANT TO CONTINUE?
> Y
RANGE OF PROBABILITIES  0.66667#-02  0.12667
    11
    1234
    ----
11/ 000.
12/ *23.
13/ 3671
14/ 3440
15/ 0000
16/ .000
ENTER TOTAL EFFORT AND DETECTION RATE
>1.0001.000,
MYOPIC PLAN?
>n
ENTER DISTRIBUTION OF SEARCH EFFORT
>11120.7
>13130.3
>0
AFTER 2 PERIODS, PROB DET IS 0.3569
WANT TO CONTINUE
>n
WANT TO PLAY AGAIN?
>n
R; &=7.10/8.93 14.06.39

```


PROGRAM SRCH1

```

CCMCN CELL(100,100),EF(100,100),PAT(60),
2 MAP(60,2),LIM(4),NAT(4),
3 MAXX,MINX,MAXY,MINY
DIMENSION NIMB(4)
DATA LAE,LIB,LOB,LY,3HREF,3HABS/
FORMAT(2X,'ENTER LIMITS OF TARGET AREA AND TYPE OF EOUNCARIES')
FCRMAIT(2X,'ENTER LIMITS OF SEARCH AREA')
1011 FCRMAIT(4I3)
101 FCRMAIT(4(I3,A3))
102 FCRMAIT(2X,'ENTER TRANSITION MATRI')
103 FCRMAIT(2I3,2F10.8)
104 FCRMAIT(2X,'INVALID INPUT DATA')
105 FCRMAIT(2X,'ENTER A PRIORI DISTRIBUTION OF TARGET')
105C FCRMAIT(2X,'ENTER TOTAL EFFCRT AND DETECTION RATE')
1051 FCRMAIT(F5.3,F5.3)
1052 FCRMAIT(2X,'MYOPIC PLAN?')
106 FCRMAIT(2X,'ENTER DISTRIBUTION OF SEARCH EFFORT')
108 FCRMAIT(2X,'WANT TC CCNTINUE?')
109 FCRMAIT(A1)
111 FCRMAIT(2I3,'I2,4X','I2','I2','I2','I2')
112 FCRMAIT(2X,'WANT TO PLAY AGAIN?')
120 WRITE(6,100)
122 READ(5,101)((LIM(I),NAT(I)),I=1,4)
IF((LIM(1).GE.1).AND.(LIM(2).LE.100).AND.(NIMB(3).GE.1).
2 (LIM(3).GE.1).AND.(LIM(4).LE.100))
3 GO TO 1020
WRITE(6,104)
GC TC 120
1020 WRITE(6,1010) (NIMB(J),J=1,4)
READ(5,1011) (NIMB(I),I=1,4)
IF((NIMB(1).GE.1).AND.(NIMB(2).LE.100).AND.(NIMB(3).GE.1).
2 (NIMB(4).LE.100)) GO TO 123
WRITE(6,104)
GC TO 1020
123 TCAL=0
DC 124 I=1,60
MAP(I,1)=0
MAP(I,2)=0
PAT(I)=0.
CCNTINUE
124 IT=1
WRITE(6,102)
125 READ(5,103) MAP(IT,1),MAP(IT,2),PAT(IT)
IF((MAP(IT,1).EQ.0).AND.(MAP(IT,2).EQ.0).AND.
2 (PAT(IT).EQ.0)) GC TO 127
TCAL=TCAL+PAT(IT)
I1=IT+1
IF(IT.LE.60) GO TO 125

```

SEA00010
SEA00020
SEA00030
SEA00040
SEA00050
SEA00060
SEA00070
SEA00080
SEA00090
SEA00100
SEA00110
SEA00120
SEA00130
SEA00140
SEA00150
SEA00160
SEA00170
SEA00180
SEA00190
SEA00200
SEA00210
SEA00220
SEA00230
SEA00240
SEA00250
SEA00260
SEA00270
SEA00280
SEA00290
SEA00300
SEA00310
SEA00320
SEA00330
SEA00340
SEA00350
SEA00360
SEA00370
SEA00380
SEA00390
SEA00400
SEA00410
SEA00420
SEA00430
SEA00440
SEA00450
SEA00460
SEA00470
SEA00480


```

127 IF (ABS(TOTAL-1.)).LE.0.0001) GO TO 129
    WRITE (6,104)
    GC TO 123
128 CONTINUE
132 TOTAL=0.
    MAXX=0
    MINX=100
    MAXY=0
    MINY=100
    DC 127 I=1,100
    DC 137 J=1,100
    CELL(I,J)=0.
    CONTINUE
137 WRITE (6,105)
141 READ (5,103) N,M,CELL(M,N)
    IF ((N.EQ.0).AND.(M.EQ.0)).AND.(CELL(M,N).EQ.0.)) GO TO 147
    MAXX=MAX0(MAXX,N)
    MINX=MIN0(MINX,N)
    MAXY=MAX0(MAXY,M)
    MINY=MIN0(MINY,M)
    TOTAL=TOTAL+CELL(M,N)
    GC TO 141
147 IF (ABS(TOTAL-1.)).LE.0.0001) GO TO 149
    WRITE (6,104)
    GC TO 133
149 KCLNT=1
155 PG=1.
    CALL TABLE (6,1050)
    WRITE (5,1051) AM,FAC
    READ (5,1051) AM,FAC
    IF (FAC.LT.1.E-5) FAC=1.E-5
    WRITE (6,1052)
    READ (5,109) IANS
    IF (IANS.NE.LAB) GO TO 159
    CALL MYOPIC(AM,FAC,NIMB)
    GC TO 181
159 DC 160 I=1,100
    DC 160 J=1,100
    EF(I,J)=0.
    CONTINUE
160 TOTAL=0.
    WRITE (6,106)
163 READ (5,103) N,M,EF(M,N)
    IF ((N.EQ.0).AND.(M.EQ.0)).AND.(EF(M,N).EQ.0.)) GO TO 171
    TOTAL=TOTAL+EF(M,N)
    GC TO 163
171 IF (ABS(TOTAL-AM)).LE.1.E-03) GO TO 181
    WRITE (6,104)

```

```

SEA00490
SEA00500
SEA00510
SEA00520
SEA00530
SEA00540
SEA00550
SEA00560
SEA00570
SEA00580
SEA00590
SEA00600
SEA00610
SEA00620
SEA00630
SEA00640
SEA00650
SEA00660
SEA00670
SEA00680
SEA00690
SEA00700
SEA00710
SEA00720
SEA00730
SEA00740
SEA00750
SEA00760
SEA00770
SEA00780
SEA00790
SEA00800
SEA00810
SEA00820
SEA00830
SEA00840
SEA00850
SEA00860
SEA00870
SEA00880
SEA00890
SEA00900
SEA00910
SEA00920
SEA00930
SEA00940
SEA00950
SEA00960

```


SEAC097C
 SEAD0980
 SEAC099C
 SEAD01000
 SEAC101C
 SEAD01020
 SEAC1030
 SEAC1040
 SEAD01050
 SEAC106C
 SEAD01070
 SEAC1080
 SEAD01090

```

GC TO 155
181 CALL PROB (PG,KOUNT,FAC)
    WRITE (6,108)
    READ (5,109) IANS
    IF (IANS.NE.LAB) GC TO 189
    CALL UPDATE (LIB,LCB,IT,FAC)
    KOUNT=KOUNT+1
    GO TO 155
185 WRITE (6,112) IANS
    REAC (5,109) IANS STOP
    IF (IANS.NE.LAB)
GC TO 120
END

```


SEAO1100
SEAO1110
SEAO1120
SEAO1130
SEAO1140
SEAO1150
SEAO1160
SEAO1170
SEAO1180
SEAO1190
SEAO1200
SEAO1210
SEAO1220
SEAO1230
SEAO1240
SEAO1250
SEAO1260
SEAO1270
SEAO1280
SEAO1290
SEAO1300
SEAO1310
SEAO1320
SEAO1330
SEAO1340
SEAO1350
SEAO1360
SEAO1370
SEAO1380
SEAO1390
SEAO1400
SEAO1410
SEAO1420
SEAO1430
SEAO1440
SEAO1450
SEAO1460
SEAO1470
SEAO1480
SEAO1490
SEAO1500
SEAO1510
SEAO1520
SEAO1530
SEAO1540
SEAO1550
SEAO1560
SEAO1570

```

SLEROUT INE MYOPIC (A,FAC,NIMB)
COMMON CELL(100,100),EF(100,100),FAT(60),
2 MAP(60,2),LIM(4),NAT(4),
3 MAXX,MINX,MAXY,MINY
DIMENSION ICN ICELL(100,100),NIMB(4)
DATA INDIA/IHY/
FORMAT (2X,'WANT TC KNOW DISTRIBUION OF EFFCRT?')
101 FORMAT (A1)
102 FORMAT (2X,' CELL EFFORT ')
103 FCRMAT (2X,I3,I3,I3,2X,F6.4)
104 FCRMAT (2X,I=MINY,MAXY
200 DC 201 I=MINY,MAXY
DC 201 J=MINX,MAXX
ICELL(I,J)=0
ICNTINUE
201 DC 231 I=MINY,MAXY
226 DC 231 J=MINX,MAXX
EF(I,J)=0.
231 CCNTINUE
241 IMIN=0
JMIN=0
B=1.
N=C
PMIN=1.
IFLAG=0
LC 247 J=MINX,MAXX
DC 247 J=MINX,MAXX
IF ((CELL(I,J).EQ.0.)OR.(ICELL(I,J).EQ.1).OR.(I.LT.NIMB(3)).OR.
2 (I.GT.NIMB(4)).OR.(J.LT.NIMB(1)).OR.(J.GT.NIMB(2))))
3 GO TO 247
IF ((CELL(I,J).GE.FMIN).OR.(IFLAG.EQ.1))GO TO 243
PMIN=CELL(I,J)
IMIN=I
JMIN=J
CCNTINUE
243 IF ((CELL(I,J).LE.PMIN).AND.(IFLAG.EQ.1)) ICELL(I,J)=1
IF (IFLAG.EQ.1) GO TO 247
N=N+1
B=(B*(1./(FAC*FLOAT(N))))*((FAC*FLOAT(N-1)))
B=R*CELL(I,J)*(1./(FAC*FLOAT(N)))
REF=EXP(A/FLCAT(N))
REF=B/(PMIN*(1./FAC))
IF (REF.GE.BEF) GO TO 247
ICELL(IMIN,JMIN)=1
IFLAG=1
CCNTINUE
247 IF (IFLAG.EQ.0) GO TO 248
GC TO 241
248 DC 251 I=MINY,MAXY

```



```

DC 251 J=MINX,MAXX
IF ((CELL(I,J).EC.C.).OR.(ICELL(I,J).NE.O)).CR.
  2 (I.LT.NIMB(3)).OR.(I.GT.NIMB(4)).OR.(J.LT.NIMB(1)).OR.
  3 (J.GT.NIMB(2))) GO TO 251
FFCB=B*((FAC*FLOAT(N))
EF(I,J)=(ALCG(CELL(I,J))/FAC)+(A/LOAT(N))-
  2 (ALCG(PROB))/(FAC*FLOAT(N)))
251 CONTINUE
WRITE (6,101)
READ (5,102) IPU
IF (IPU.NE.INDIA) RETURN
WRITE (6,103)
DC 281 I=MINY,MAXY
DO 281 J=MINX,MAXX
IF (EF(I,J).EC.O.) GO TO 281
WRITE (6,104) J,I,EF(I,J)
281 CONTINUE
RETURN
END

```

```

SEA01580
SEA01550
SEA01600
SEA01610
SEA01620
SEA01630
SEA01640
SEA01650
SEA01660
SEA01670
SEA01680
SEA01690
SEA01700
SEA01720
SEA01730
SEA01740
SEA01750
SEA01760

```



```

SUBROUTINE UPDATE (LIB,LOB,IT,FAC)
COMMON CELL(100,100),EF(100,100),PAT(60),
2 MAP(60,2),LIM(4),NAT(4),
3 MAXX,MINX,MAXY,MINY
DENOM=0
DC 318 I=1,100
DC 318 J=1,100
DENOM=DENOM+CELL(I,J)*EXP(-FAC*EF(I,J))
318 CCNT=CCNT+1
DO 319 I=1,100
DO 319 J=1,100
CELL(I,J)=CELL(I,J)*EXP(-FAC*EF(I,J))/DENOM
EF(I,J)=0.
319 CCNT=CCNT+1
CALL SPREAD (LIB,LOB,IT)
RETURN
END

```

```

SEA02230
SEA02240
SEA02250
SEA02260
SEA02270
SEA02280
SEA02290
SEA02300
SEA02310
SEA02320
SEA02330
SEA02340
SEA02350
SEA02360
SEA02370
SEA02380
SEA02390

```


AC240C
AC2410
AC2420
AC2430
AC2440
AC2450
AC2460
AC2470
AC2480
AC2490
AC2500
AC2510
AC2520
AC2530
AC2540
AC2550
AC2560
AC2570
AC2580
AC2590
AC2600
AC2610
AC2620
AC2630
AC2640
AC2650
AC2660
AC2670
AC2680
AC2690
AC2700
AC2710
AC2720
AC2730
AC2740
AC2750
AC2760
AC2770
AC2780
AC2790
AC2800
AC2810

```

SLROUTINE SPREAD (LIE,LOB,IT)
COMMON CELL(100,100),EF(100,100),PAT(60),
MAP(60,2),LIM(4),NAT(4),
MAXX,MINX,MAXY,MINY
MAXX=0
MINX=100
MAXY=0
MINY=100
DO 419 I=1,100
  DO 418 J=1,100
    IF (CELL(J,I).EQ.0.) GO TO 415
    DG 418 N=1,IT
    LI=I+MAP(N,1)
    LJ=J+MAP(N,2)
    IF ((LI.LT.LIM(1)).AND.(LI.GE.LIM(1)).AND.(LI.GE.LIM(1)).AND.(NAT(1).EQ.LCB)).OR.
      ((LI.GT.LIM(2)).AND.(LI.LE.LIM(2)).AND.(NAT(2).EQ.LOB)).CR.
      ((LIM(1).LE.I).AND.(LIM(2).GE.I)).AND.
      ((LI.LT.LIM(1)).AND.(NAT(1).EQ.LIB)).OR.
      ((LI.GT.LIM(2)).AND.(NAT(2).EQ.LIB))))
      LI=I-MAP(N,1)
    IF ((LI.LT.LIM(3)).AND.(LI.GE.LIM(3)).AND.(NAT(3).EQ.LOB)).OR.
      ((LI.GT.LIM(4)).AND.(LI.LE.LIM(4)).AND.(NAT(4).EQ.LOB)).OR.
      ((LIM(3).LE.J).AND.(LIM(4).GE.J)).AND.
      ((LJ.LT.LIM(3)).AND.(NAT(3).EQ.LIB)).OR.
      ((LJ.GT.LIM(4)).AND.(NAT(4).EQ.LIB))))
      LJ=J-MAP(N,2)
    IF ((LI.LT.1).OR.(LI.GT.100).OR.(LJ.LT.1).OR.(LJ.GT.100))
      GO TO 415
    EF (LJ,LI)=EF(LJ,LI)+PAT(N)*CELL(J,I)
    CCNTINUE
  DO 518 I=1,100
    CC 518 J=1,100
    CELL(I,J)=EF(I,J)
    IF (CELL(I,J).EQ.0.) GO TO 518
    MAXX=MAX0(MAXX,J)
    MINX=MIN0(MINX,J)
    MAXY=MAX0(MAXY,I)
    MINY=MIN0(MINY,I)
  CCNTINUE
518 RETURN
END

```



```

SLROUTINE FROB(PG,KOUNT,FAC)
COMMON CELL(100,100),EF(100,100),PAT(60),
2 MAP(60,2),LIM(4),NAT(4),
3 MAXX,MINX,MAXY,MINY
100 FORMAT(2X,'AFTER',1X,12,1X,'PERIODS',PROE[ET IS ',G12.5)
PNOT=0. I=MINY,MAXY
DO 517 J=MINX,MAXX
PNOT=PNOT+CELL(I,J)*(1.-EXP(-FAC*EF(I,J)))
517 CCNTINUE
PNOT=1.-PNOT
FG=PG*PNOT
PNOT=1.-PG
WRITE(6,100) KOUNT,FNOT
RETURN
ENC

```

```

AC282C
AC2830
AC2840
AC2850
AC2860
AC2870
AC2880
AC2890
AC290C
AC2910
AC292C
AC2930
AC2940
AC2950
AC2960
AC297C

```


PROGRAM SRCH2

```

COMMON CELL(25,25,4,4),EF(25,25),PDSX(10),PCSY(10)
2 NSX(4),NSY(4),IDSX(10),IDSY(10),LIM(4),NAT(4),
3 MAXX,MINX,MAXY,MINY
DIMENSION NIMB(4)
DATA LAB,LIB,LOB/1HY,3HREF,3HABS/
100 FCRMAT(2X,'ENTER LIMITS OF TARGET AREA AND TYPES OF BOUNCARIES')
91 FCRMAT(4(I2,A3))
51 FCRMAT(2X,'ENTER LIMITS OF SEARCH AREA')
100 FCRMAT(4(I2))
100 FCRMAT(2X,'HOW MANY VALUES CAN V(X) ASSUME?')
101 FCRMAT(2X,'INVALID INPUT DATA')
102 FCRMAT(2X,'HOW MANY VALUES CAN V(Y) ASSUME?')
103 FCRMAT(2X,'HOW MANY VALUES CAN CNE PER LINE?')
104 FCRMAT(2X,'ENTER THESE VALUES: CNE PER LINE')
105 FCRMAT(2X,'ENTER TRANSITION OF V(X)')
106 FCRMAT(12,F5.3)
107 FCRMAT(2X,'TRANSITION OF V(Y)')
108 FCRMAT(2X,'ENTER A PRIORI DISTRIBUTIONS')
109 FCRMAT(2I2,11F5.3)
1090 FCRMAT(2X,'ENTER TOTAL EFFORT AND DETECTION RATE')
1091 FCRMAT(2F5.3)
1092 FCRMAT(2X,'MYOPIC PLAN?')
110 FCRMAT(2X,'ENTER DISTRIBUTION OF SEARCH EFFCRT')
111 FCRMAT(2X,'WANT TO CONTINUE?')
112 FCRMAT(2X,'REPEAT LAST ENTRY')
113 FCRMAT(2X,'WANT TO PLAY AGAIN?')
114 FCRMAT(2X,'WANT TO PLAY AGAIN?')
115 FCRMAT(2X,'WANT TO PLAY AGAIN?')
1180 FCRMAT(2X,'WANT TO PLAY AGAIN?')
WRITE(6,90)
READ(5,91) ((LIM(I),NAT(I)),I=1,4)
IF ((LIM(1).GE.1).AND.(LIM(2).LE.25).AND.(LIM(3).GE.1).
AND.(LIM(4).LE.25)) GO TO 185
2 WRITE(6,102)
GC TO 180
WRITE(6,100)
189 WRITE(5,100C) (NIMB(J),J=1,4)
READ(5,100C) (NIMB(2).LE.25).AND.(NIMB(3).
IF ((NIMB(1).GE.1).AND.(NIMB(4).LE.25)) GC TO 200
2 WRITE(6,102)
GC TO 189
200 MAXSX=-1000
MINSX=1000
MAXSY=-1000
MINSY=1000
201 WRITE(6,101)
READ(5,106) IX
IF (IX.LE.4) GO TC 202
WRITE(6,102)
GC TO 201
202 WRITE(6,104)

```



```

203 I=1,IX
204 READ (5,106) NSX(I)
      MAXSX=MAX0(MAXSX,NSX(I))
      MINSX=MIN0(MINSX,NSX(I))
      CCNTINUE
205 WRITE (6,102)
      READ (5,106) IY
      IF (IY.LE.4) GO TO 205
      WRITE (6,102)
      GO TO 204
206 WRITE (6,104)
      CC 206 I=1,IY
      READ (5,106) NSY(I)
      MAXSY=MAX0(MAXSY,NSY(I))
      MINSY=MIN0(MINSY,NSY(I))
      CCNTINUE
209 WRITE (6,105)
      DC 208 I=1,10
      ICSX(I)=0
      PCSX(I)=0
208 CCNTINUE
      EF(17,19)=0.
      INX=1
211 READ (5,106) IDSX(INX),PDSX(INX)
      IF ((IDSX(INX).EQ.0).AND.(PDSX(INX).EQ.0.)) GO TO 215
      EF(17,19)=EF(17,19)+PDSX(INX)
      INX=INX+1
      IF (INX.LE.10) GO TO 211
215 IF (ABS(EF(17,19)-1.).LE.1.E-04) GC TO 235
      WRITE (6,102)
      GO TO 209
235 INX=INX-1
      DC 236 I=1,10
      ICSY(I)=0
      PCSY(I)=0
236 CCNTINUE
      EF(21,25)=0.
      INY=1
237 READ (5,106) IDSY(INY),PDSY(INY)
      IF ((IDSY(INY).EQ.0).AND.(PDSY(INY).EQ.0.)) GO TO 239
      EF(21,25)=EF(21,25)+PDSY(INY)
      INY=INY+1
      IF (INX.LE.10) GO TO 237
239 IF (ABS(EF(21,25)-1.).LE.1.E-04) GC TO 253
      WRITE (6,102)
      GO TO 235
253 EF(3,3)=0.

```

```

004500
005000
005100
005200
005300
005400
005500
005600
005700
005800
005900
006000
006100
006200
006300
006400
006500
006600
006700
006800
006900
007000
007100
007200
007300
007400
007500
007600
007700
007800
007900
008000
008100
008200
008300
008400
008500
008600
008700
008800
008900
009000
009100
009200
009300
009400
009500
009600

```



```

INY=INY-1
DC 2660 I=1,25
DC 2660 J=1,25
DC 2660 K=1,4
DC 2660 L=1,4
CELL(I,J,K,L)=0.
CONTINUE
2660 EF(20,20)=0.
MAXX=0
MINX=100
MAXY=0
MINY=100
WRITE(6,108)
254 READ(5,109) I,J,EF(20,20), (EF(1,K),K=1,IX), (EF(2,L),L=1,IY)
IF ((I.EQ.0).AND.(J.EQ.0).AND.(EF(20,20).EQ.0.)) GO TO 262
EF(5,2)=0.
DC 259 M=I, IY
EF(5,2)=EF(5,2)+EF(2,M)
EF(5,1)=0.
DC 259 N=I, IX
EF(5,1)=EF(5,1)+EF(1,N)
CELL(J,I,M,N)=EF(20,20)*EF(2,M)*EF(1,N)
CONTINUE
259 IF ((ABS(EF(5,1)-1.)*GT.1.0E-04).OR.
1 (ABS(EF(5,2)-1.)*GT.1.0E-04)) GO TO 260
MAXX=MAX0(MAXX,I)
MINX=MIN0(MINX,I)
MAXY=MAX0(MAXY,J)
MINY=MIN0(MINY,J)
GC TO 254
260 EF(20,20)=0.
DC 261 M=1,4
EF(1,M)=0.
EF(2,M)=0.
DC 261 N=1,4
CELL(I,J,M,N)=0.
CONTINUE
261 WRITE(6,102)
16,115)
GC TO 254
262 IF (ABS(EF(3,3)-1.)*LE.1.0E-04) GC TO 264
DC 263 I=1,25
DC 263 J=1,25
DC 263 K=1,4
DC 263 L=1,4
CELL(I,J,K,L)=0.
CONTINUE
263

```



```

264 WRITE (6,102)
    GC TO 253
265 KCUNT=1
    PG=1. TABLE (IX,IY)
    CALL PR (6,1090)
    READ (5,1091) A,FAC
    IF (FAC.LT.1.E-05) FAC=1.E-05
    WRITE (6,1092)
    READ (5,115) IANS
    IF (IANS.NE.LAB) GO TO 2067
    CALL MYOPIC (A,FAC,IX,IY,NIMB)
    GC TO 281
2067 TCT=0. I=1,25
    DO 266 J=1,25
    DC 266 J=1,25
    EF(I,J)=0.
    CCNTINUE
266 WRITE (6,110) I,J,EF(J,I)
267 READ (5,109) I,J,EF(J,I)
    IF ((I.EQ.0).AND.(J.EQ.0).AND.(EF(J,I).EQ.0.)) GO TO 269
    TCT=TCT+EF(J,I)
    EF(J,I)=EF(J,I)*FAC
    GC TO 267
269 IF (ABS(TOT-A).LE.1.E-03) GO TO 281
    WRITE (6,102)
    GC TO 2067
281 CALL PROB (PG,KOUNT,IX,IY)
    WRITE (6,112)
    READ (5,115) IANS
    IF (IANS.NE.LAB) GC TO 283
    CALL UPDATE (IX,IY,INX,INY,MAXSX,MINSX,MAXSY,MINSY,LIB,LCB)
    KCUNT=KOUNT+1
    GC TO 265
283 WRITE (6,114)
    READ (5,115) IANS
    IF (IANS.NE.LAB) STOP
    GC TO 180
END

```

```

SC01450
SECC01460
SECC01470
SECC01480
SECC01490
SECC01500
SECC01510
SECC01520
SECC01530
SECC01540
SECC01550
SECC01560
SECC01570
SECC01580
SECC01590
SECC01600
SECC01610
SECC01620
SECC01630
SECC01640
SECC01650
SECC01660
SECC01670
SECC01680
SECC01690
SECC01700
SECC01710
SECC01720
SECC01730
SECC01740
SECC01750
SECC01760
SECC01770
SECC01780
SECC01790
SECC01800
SECC01810
SECC01820
SECC01830

```



```

IF (REF.GE.PEF) GC TO 247
IFLAG=1
IFLAG=1
CCNTINUE
247 IF (IFLAG.EQ.0) GO TO 248
GC TO 241
DC 251 I=MINY,MAXY
CC 251 J=MINX,MAXX
IF ((TCELL(I,J).EQ.0.)OR.(ICELL(I,J).EQ.1).OR.(I.LT.NIMB(3)).
OR.(I.GT.NIMB(4)).OR.(J.LT.NIMB(1)).OR.(J.GT.NIME(2)))
3 GO TO 251
3 PROB=B*((FAC*FLOAT(N))
EF(I,J)=(ALOG(TCELL(I,J))/FAC)+(A/LOAT(N))-
2
251 CONTINUE
WRITE(6,101) IPU
READ(5,102) INDIA) RETURN
IF (IPU.NE.INDIA)
WRITE(6,103)
DC 281 I=MINY,MAXY
DC 281 J=MINX,MAXX
IF (EF(I,J).EQ.0.) GO TO 281
WRITE(6,104) J,I,EF(I,J)
281 CONTINUE
RETURN
END

```



```

SLBROUT INE TABLE (IX,IY)
COMMON CELL(25,25,4,4),EF(25,25),PDSX(10),PCSY(10),
2 NSX(4),NSY(4),IDSX(10),IDSY(10),LIM(4),NAT(4),
3 MAXX,MINX,MAXY,MINY
DIMENSION ICCELL(25),IMBOL(15)
DATA IMBOL/1H0,1H1,1H2,1H3,1H4,1H5,1H6,1H7,1H8,1H9,
2 1H*,1H.,1H-,1H/,
FCRMAT (2X,I3,.,/,100A1)
100 FCRMAT (2X,I3,.,/,100A1)
101 FCRMAT (6X,I2,9(8X,I2))
102 FCRMAT (7X,100I1)
103 FCRMAT (7X,100I1)
104 FCRMAT (7X,100A1)
DC 201 I=1,25
DC 201 J=1,25
EF(I,J)=0.
DC 201 K=1,IY
DC 201 L=1,IX
EF(I,J)=EF(I,J)+CELL(I,J,K,L)
201 CCNTINUE
SMALL=1.
BIG=0.
DC 217 I=MINY,MAXY
DC 217 J=MINX,MAXX
IF ((EF(I,J)).NE.0.).AND.(EF(I,J).LT.SMALL)) SMALL=EF(I,J)
IF ((EF(I,J)).GT.BIG) BIG=EF(I,J)
CCNTINUE
217 WRITE (6,100) SMALL,BIG
IF ((BIG-SMALL).LT.I.E-5) SMALL=0.
DC 417 I=1,25
ICELL(I)=MINX+(I-1)*10
IF (ICELL(I).GT.MAXX) GO TO 418
CCNTINUE
417 I=I-1
418 WRITE (6,102) (ICELL(J),J=1,I)
LA=MAXX-MINX+1
DC 517 N=1,LA
ICELL(N)=MCC(N-1+MINX,10)
CCNTINUE
517 WRITE (6,103) (ICELL(I),I=1,LA)
DC 617 I=1,LA
ICELL(I)=IMBOL(13)
CCNTINUE
617 WRITE (6,104) (ICELL(I),I=1,LA)
DC 817 I=MINY,MAXY
DC 717 J=MINX,MAXX
ICELL(J)=IMBOL(1)+
2 IFIX(10.-SQRT(100.-(EF(I,J)-SMALL)/(BIG-SMALL)*100.)))
IF (EF(I,J).EQ.0.) ICCELL(J)=IMBOL(12)

```

```

SEC02580
SEC02590
SEC02600
SEC02610
SEC02620
SEC02630
SEC02640
SEC02650
SEC02660
SEC02670
SEC02680
SEC02690
SEC02700
SEC02710
SEC02720
SEC02730
SEC02740
SEC02750
SEC02760
SEC02770
SEC02780
SEC02790
SEC02800
SEC02810
SEC02820
SEC02830
SEC02840
SEC02850
SEC02860
SEC02870
SEC02880
SEC02890
SEC02900
SEC02910
SEC02920
SEC02930
SEC02940
SEC02950
SEC02960
SEC02970
SEC02980
SEC02990
SEC03000
SEC03010
SEC03020
SEC03030
SEC03040
SEC03050

```



```
717 CCNTINUE (6,101) I,(ICELL(J),J=MINX,MAXX)
817 WRITE (6,101) I,(ICELL(J),J=MINX,MAXX)
817 CCNTINUE
817 RETURN
817 END
```

```
SECO3060
SECO3070
SECO3080
SECO3090
SECO3100
```



```

SLBROUTINE PROB( PG,KCUNT,IX,IY)
CCCOMMON CELL(25,25,4,4),EF(25,25),PDSX(10),PCSY(10),
. NSX(4),NSY(4),IDSX(10),IDSY(10),LIM(4),NAT(4),
. MAXX,MINX,MAXY,MINY
101 FORMAT (2X,'AFTER',1X,12,1X,'PERICDS, PROBE LET IS ',F6.4)
PNOT=1.
DC 219 I=1,25
DC 219 J=1,25
A=0.
DC 217 K=1,IY
DC 217 L=1,IX
A=A+CELL(I,J,K,L)
217 CCONTINUE
PNOT=PNOT-A*(1.-EXP(-EF(I,J)))
219 CCONTINUE
PG=PG*PNOT
PNOT=1.-PG
WRITE(6,101) KCUNT,PNOT
RETURN
END

```

```

CC03110
CC031120
CC031130
CC031140
CC031150
CC031160
CC031170
CC031180
CC031190
CC03200
CC03210
CC03220
CC03230
CC03240
CC03250
CC03260
CC03270
CC03280
CC03290
CC03300

```



```

      SLBRoutine Lpdate (IX,IY,INX,INY,MAXSX,MINSX,MAXSY,MINSY,
      2      LIB,LOB)
      COMMON CELL(25,25,4,4),EF(25,25),PCSX(10),PCSY(10),
      2      NSX(4),NSY(4),IDSX(10),IDSY(10),LIM(4),NAT(4),
      3      MAXX,MINX,MAXY,MINY
      DIMENSION TCELL(25,25),PVY(10),FVX(10)
      DENOM=0.
      CC 319 I=1,25
      DC 319 J=1,25
      TCELL(I,J)=0.
      CC 318 K=1,IY
      DC 318 L=1,IX
      TCELL(I,J)=TCELL(I,J)+CELL(I,J,K,L)
      318 CONTINUE
      DENOM=DENOM+TCELL(I,J)*EXP(-EF(I,J))
      319 CONTINUE
      CC 351 I=1,25
      DC 351 J=1,25
      CC 339 L=1,IX
      PVX(L)=0.
      CC 338 K=1,IY
      PVX(L)=PVX(L)+CELL(I,J,K,L)
      338 CONTINUE
      IF (TCELL(I,J).EQ.0.) GO TO 339
      PVX(L)=PVX(L)/TCELL(I,J)
      339 CONTINUE
      CC 345 K=1,IY
      PVY(K)=0.
      CC 341 L=1,IX
      PVY(K)=PVY(K)+CELL(I,J,K,L)
      341 CONTINUE
      IF (TCELL(I,J).EQ.0.) GO TO 345
      PVY(K)=PVY(K)/TCELL(I,J)
      345 CONTINUE
      TCELL(I,J)=TCELL(I,J)*EXP(-EF(I,J))/DENOM
      CC 347 K=1,IY
      DC 347 L=1,IX
      CELL(I,J,K,L)=TCELL(I,J)*PVY(K)*FVX(L)
      347 CONTINUE
      351 CONTINUE
      CALL SPREAD (IX,IY,INX,INY,MAXSX,MINSX,MAXSY,MINSY,LIB,LOB)
      RETURN
      END

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SEC033310
SEC033320
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SEC033660
SEC033670
SEC033680
SEC033690
SEC033700
SEC033710
SEC033720
SEC033730

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SUBROUTINE SPREAD (IX,IY,INX,INY,MAXSX,MINSX,MAXSY,MINSY,
1      LIB,LOB)
2 COMMON CELL(25,25,4,4),EF(25,25),PCSX(10),PCSY(10),
2      NSX(4),NSY(4),IDSX(10),IDSY(10),LIM(4),NAT(4),
3      MAXX,MINX,MAXY,MINY
DIMENSION TCELL(25,25,4,4),INAX(10,4),INAY(10,4)
DC 201 I=1,25
DO 201 J=1,25
DO 201 K=1,IY
DO 201 L=1,IX
TCELL(I,J,K,L)=0.
201 CCNTINUE
DC 231 I=1,25
CC 231 J=1,25
CC 231 K=1,IY
CC 231 L=1,IX
LI=I+NSX(K)
LJ=J+NSY(K)
IF ((LI.LT.LIM(1)).AND.(LI.GE.LIM(1)).AND.(NAT(1).EQ.LOB)).OR.
2 ((LI.GT.LIM(2)).AND.(LI.LE.LIM(2)).AND.(NAT(2).EQ.LOB)).OR.
3 ((LI.LT.LIM(1)).AND.(LIM(2).GE.I)).AND.(NAT(1).EQ.LOB)).OR.
4 ((LI.LT.LIM(1)).AND.(NAT(1).EQ.LIB)).OP.
5 ((LI.GT.LIM(2)).AND.(NAT(2).EQ.LIB)))
6 LI=I-NSX(L)
IF ((LI.LT.LIM(3)).AND.(LJ.GE.LIM(3)).AND.(NAT(3).EQ.LOB)).OR.
2 ((LI.GT.LIM(4)).AND.(LJ.LE.LIM(4)).AND.(NAT(4).EQ.LOB)).OR.
3 ((LIM(3).LE.J)).AND.(LIM(4).GE.J)).AND.(NAT(3).EQ.LOB)).OR.
4 ((LJ.LT.LIM(3)).AND.(NAT(3).EQ.LIB)).CF.
5 ((LJ.GT.LIM(4)).AND.(NAT(4).EQ.LIB)))
6 LJ=J-NSY(K)
IF ((LI.LT.1).OR.(LJ.LT.1).OR.
2 ((LI.GT.25).OR.(LJ.GT.25)) GO TC 231
TCELL(LJ,LI,K,L)=TCELL(LJ,LI,K,L)+CELL(J,I,K,L)
231 CCNTINUE
MAXX=0
MINX=30
MAXY=0
MINY=30
DO 251 I=1,25
DC 251 J=1,25
DO 251 K=1,IX
DO 251 L=1,IX
IF (TCELL(I,J,K,L).EQ.0.) GO TO 245
MAXX=MAX0(MAXX,J)
MINX=MIN0(MINX,J)
MAXY=MAX0(MAXY,I)
MINY=MIN0(MINY,I)
245 CELL(I,J,K,L)=0.

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SEC03740
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SEC03990
SEC04000
SEC04010
SEC04020
SEC04030
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SEC04090
SEC04100
SEC04110
SEC04120
SEC04130
SEC04140
SEC04150
SEC04160
SEC04170
SEC04180
SEC04190
SEC04200
SEC04210
SEC04220

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251 CCNTINUE
    CC 701 I=1, INX
    CC 701 J=1, IY
    INAX(I, J)=MAXO(MINSX, MINO(MAXSX, NSX(J)+IDSX(I)))
701 CCNTINUE
    CC 702 I=1, INY
    CC 702 J=1, IY
    INAY(I, J)=MAXO(MINSY, MINO(MAXSY, NSY(J)+IDSY(I)))
702 CCNTINUE
    CC 710 I=MINY, MAXY
    CC 710 J=MINX, MAXX
    CC 710 IIX=1, INX
    CC 710 IIX=1, IY
    CC 705 N=1, IX
    IF (INAX(IIX, IIX).EQ.NSX(N)) GO TC 706
705 CCNTINUE
706 CC 710 IINY=1, INY
    CC 710 IYY=1, IY
    CC 707 M=1, IY
    IF (INAY(IINY, IYY).EQ.NSY(M)) GO TC 708
707 CCNTINUE
708 CELL(I, J, M, N)=CELL(I, J, IY, IIX)*
    PDSY(IINY)*PDSX(IIX)
710 CCNTINUE
    RETURN
    END

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SEC04220
SEC04230
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SEC04450
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SEC04470

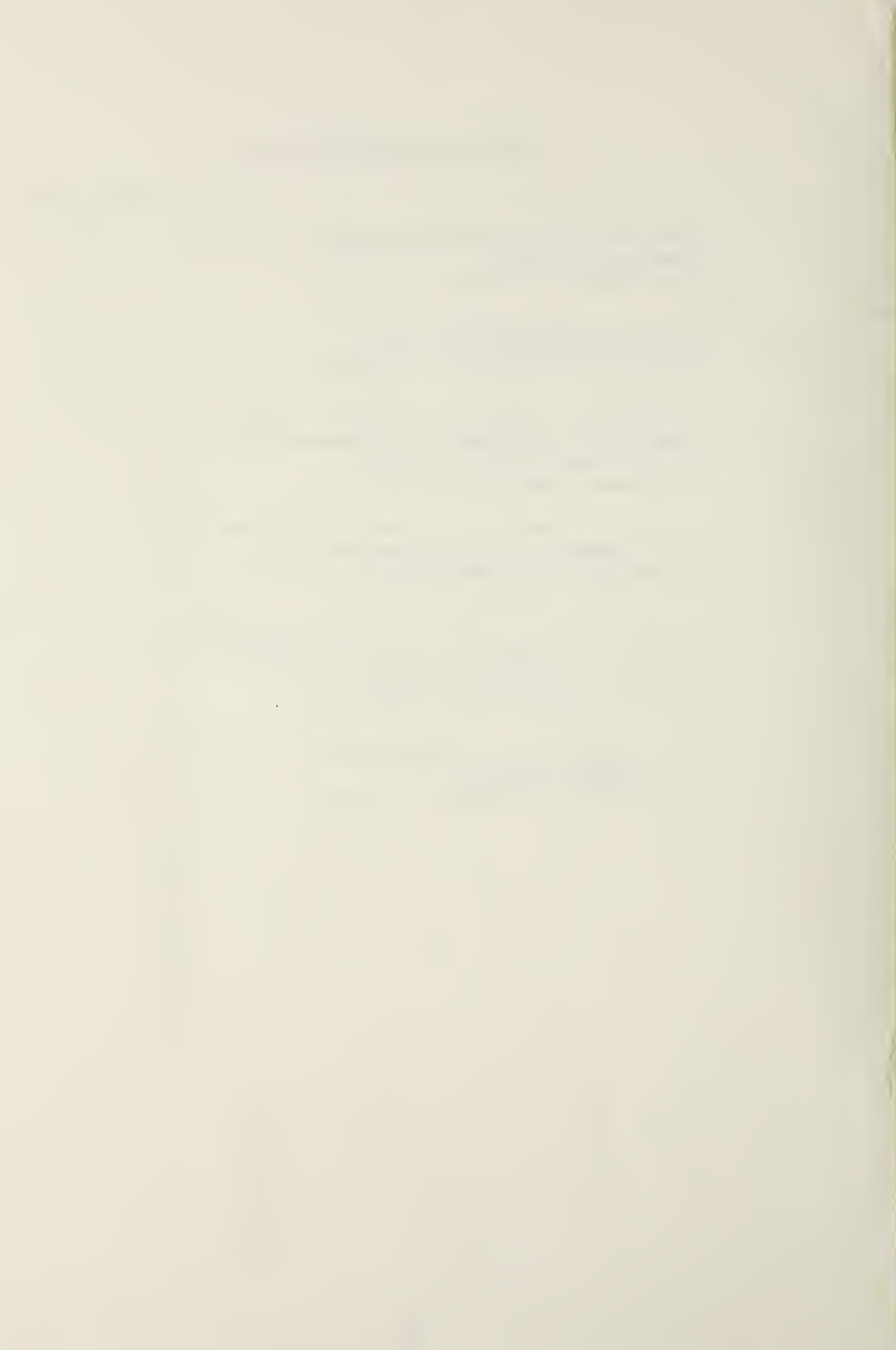
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